



Jet Propulsion Laboratory
California Institute of Technology

Kigen: Supporting Spacecraft Design through Applied Artificial Intelligence

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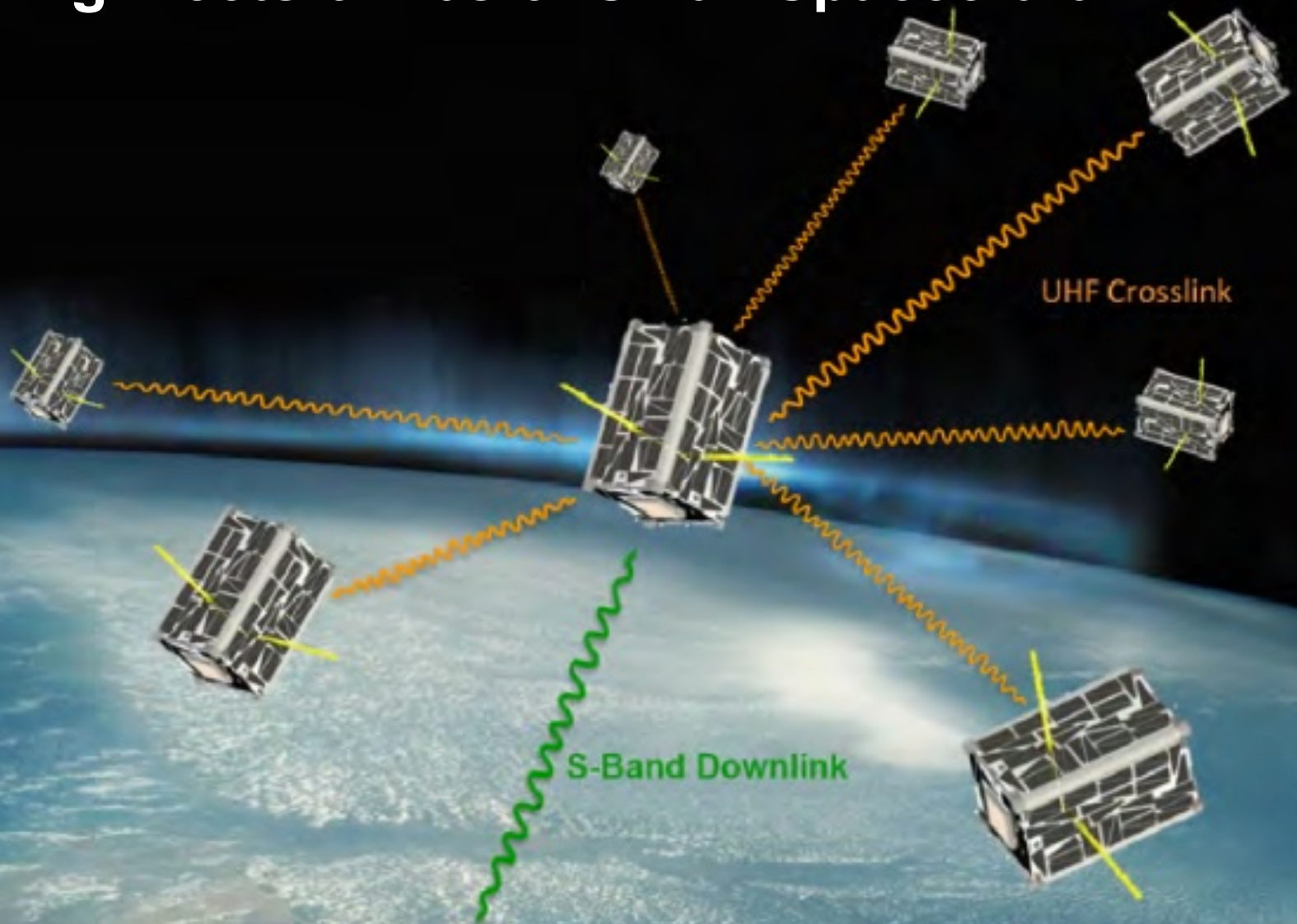
1 July 2018

Designing with AI, Design Computing and Cognition Conference (DCC'18), Milan, Italy

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NASA Scientists are Envisioning Missions Involving Fleets of 10s of Small Spacecraft...



Artist's Impression

**... to Enable New Science and
More Efficient Exploration**



**How do we best support the design of such
missions?**

Artist's Impression

Motivating Example

Spacecraft-Based Radio Interferometry



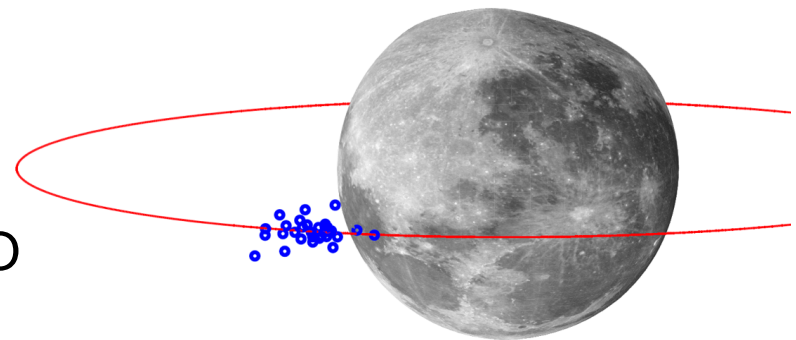
Source: <http://www.passmyexams.co.uk/GCSE/physics/images/radio-telescopes-outdoors.jpg>

Radio interferometers:

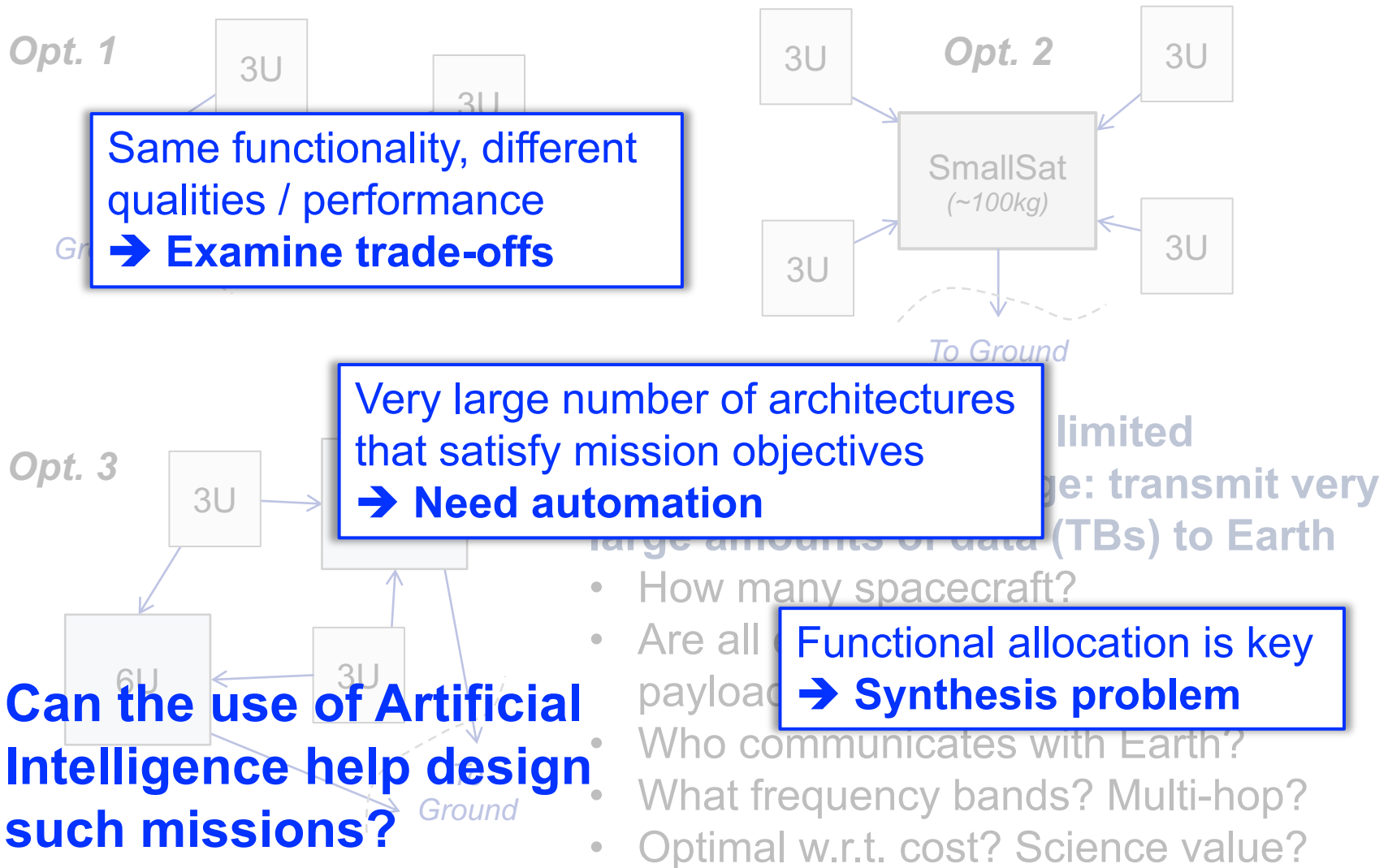
- Radio telescopes consisting of multiple low-frequency antennas
 - Achieve the same angular resolution as that of a single telescope with the same aperture
- ➔ Typically ground-based

Want to do this in space:

- Frequencies $< 30\text{MHz}$ blocked by ionosphere
 - Cluster of spacecraft (3 – 50) functioning as interferometer in LLO
- ➔ CubeSats or SmallSats are promising enablers for this



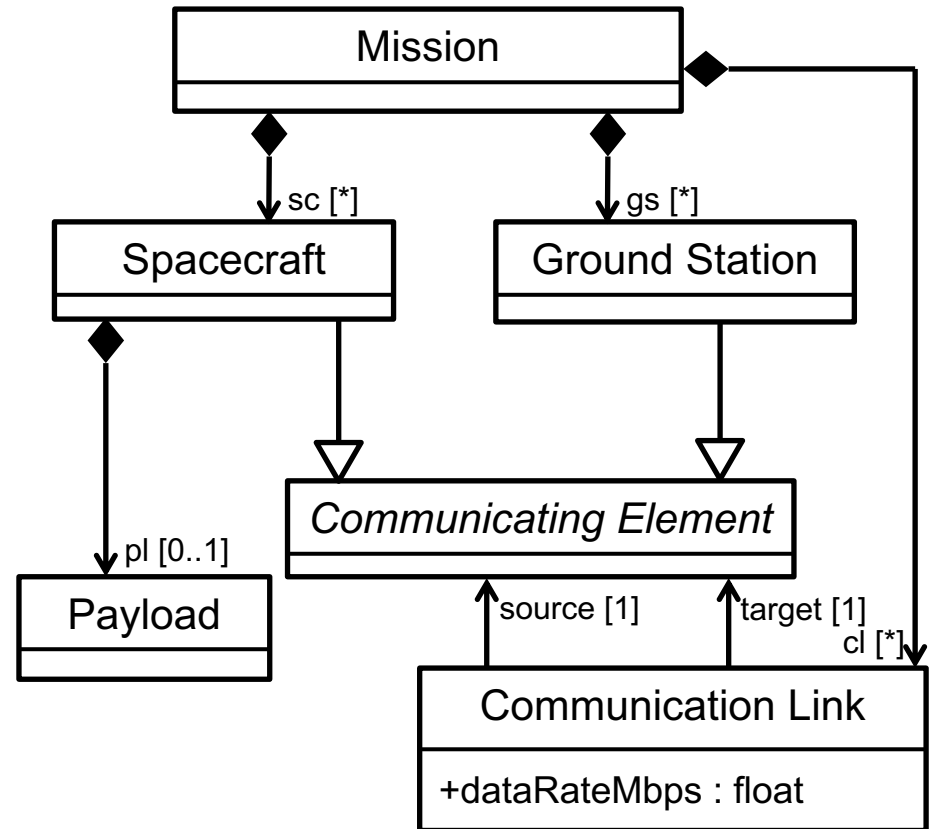
Which Architecture is Optimal?



Abstracting the Problem Domain

- Domain model
 - Concepts and behavior
 - Attributes and parametric relations
 - Associations & other relations
 - ➔ Describes a **universe of discourse**: many models in domain
 - ➔ Describes structural part of the problem
- Typically with addl. well-formedness constraints, e.g.:

“All spacecraft must (transitively) be connected to at least one ground station through a communication link”

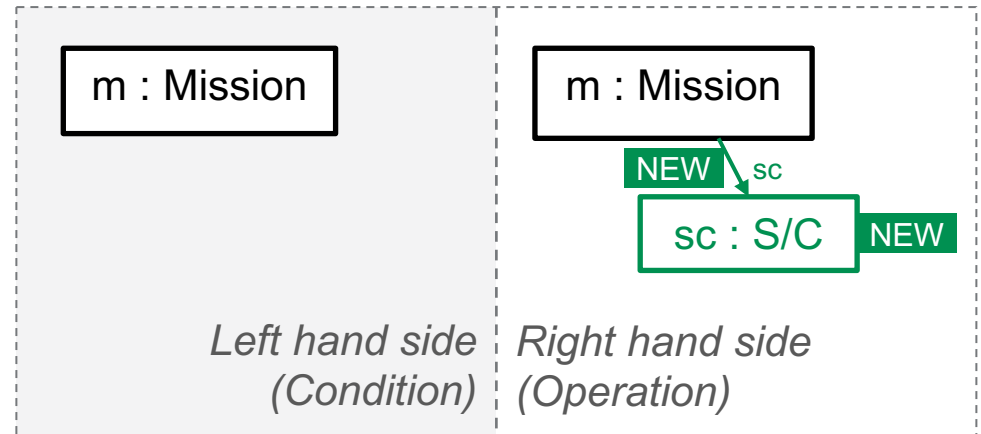


Any well-formed model
in the domain is a
candidate solution

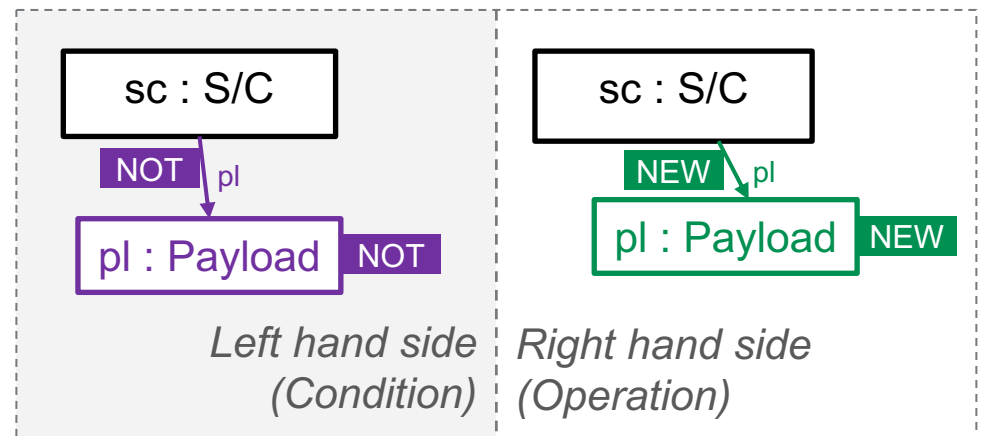
Rule-Based Model Generation

Model Transformation Rules as Enablers for Evolving Solutions

- Model transformation rules as operations for creating and / or **evolving** a given design / model in domain
 - LHS:** **Condition** for match in input model (e.g., “*find an element of type Mission*”)
 - RHS:** **Operation** to be performed (object creation, modification, deletion)
- Here: *endogenous, in-place* transformations
- Can also encode design heuristics this way!

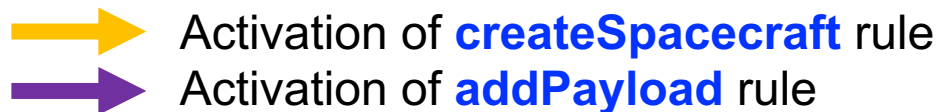


Rule “createSpacecraft”



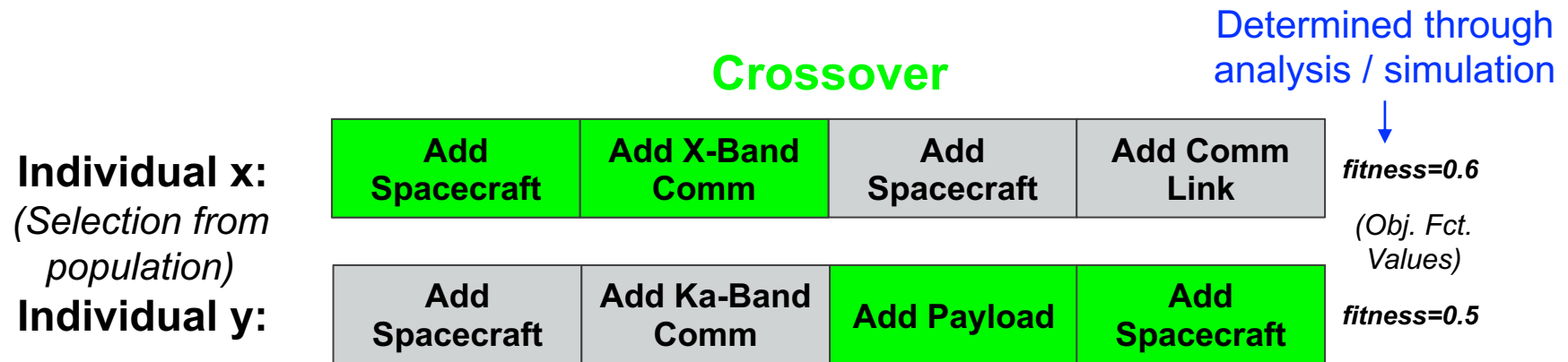
Rule “addPayload”

Forming the Model State Space



Evolving a Population of Models

Example: Using Genetic Algorithms to find Pareto-Optimal Solutions



Here, individuals are **sequences of transformation rule activations**

→ Each genome in population is a variable with set of trafo rules as range

New:

(Recombined individual in next generation)

Add Ka-Band Comm

fitness=0.9

Mutation

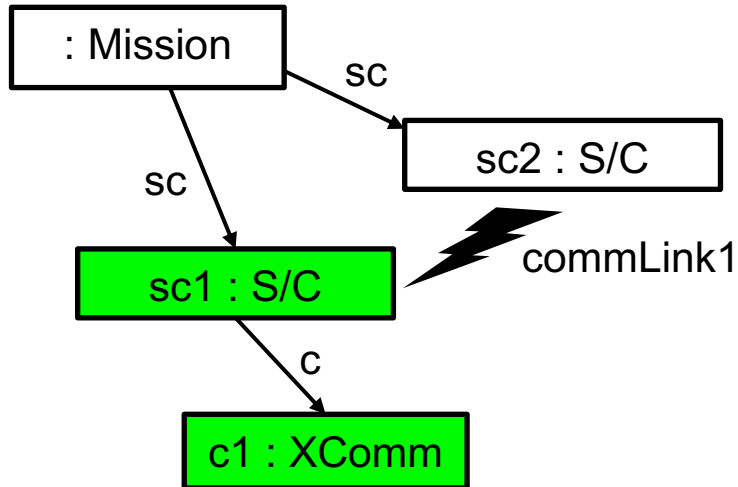
Could also be a “placeholder” transformation
(= rule “do nothing”)

Can also use other Optimization Techniques, e.g.: Hill Climbing

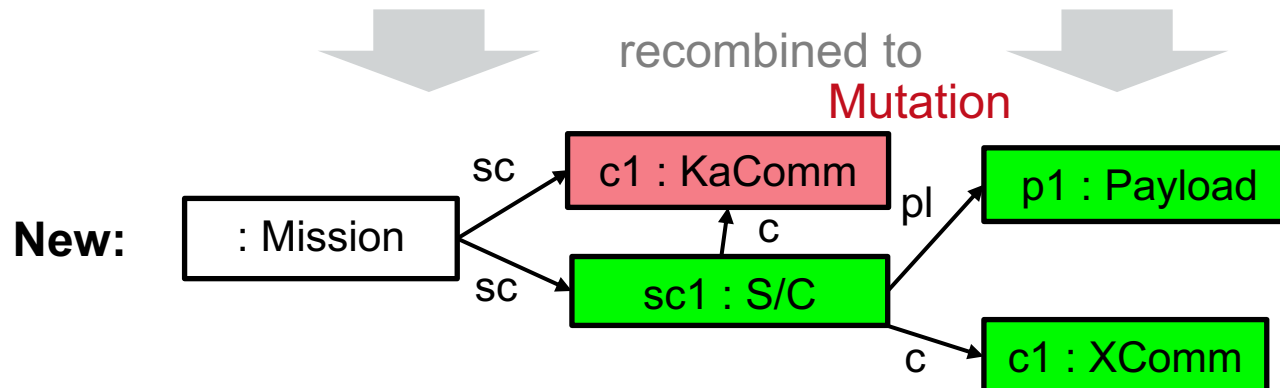
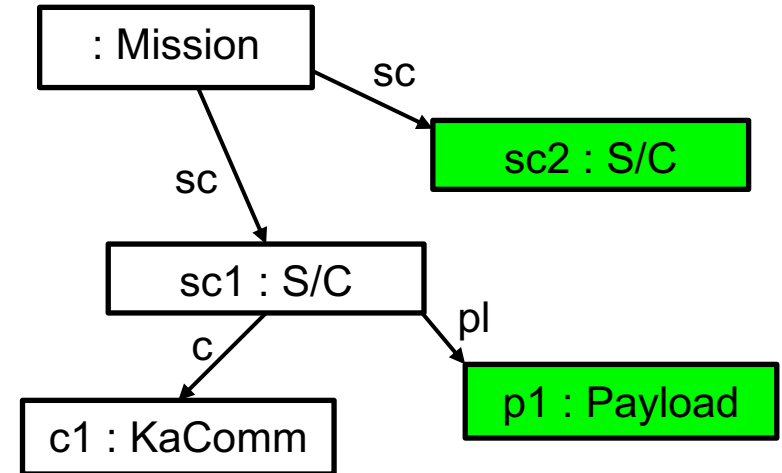
Evolving a Population of Models

Illustrative Example

Individual x:



Individual y:



The Kigen Modeling Language (KigenML)

Design Problem Definition in KigenML

```
/* Domain description */
type ConstellationMission {
  var spacecraft : Spacecraft [ 2 .. 50 ]

  fun cost = { ...// Some complex function
  fun benefit = { ... // Another function

  constraint atLeast2SC: filter(
    spacecraft,
    (s : Spacecraft)
      => s.payload != null
    ).size > 1)
}

type Spacecraft {
  var payload : Payload [ 0 .. 1 ]
}

abstract type Payload

type InterferometryPayload
  extends Payload
```

```
/* Initial conditions */
model myMission of ConstellationMission {
  model sc1 of Spacecraft in spacecraft
  model sc2 of Spacecraft in spacecraft {
    model p1 of InterferometryPayload
      in payload
  }
}

/* Define optimization criteria */
minimize myMission.cost
maximize myMission.benefit

solve optimize using 'nsgaii' 30 times
  with
    populationSize 100,
    maxGenerations 20,
    maxSolutionLength 15
```

**Able to derive domain model, rules
& compile executable optimization
problem from a Kigen program**

Implementation

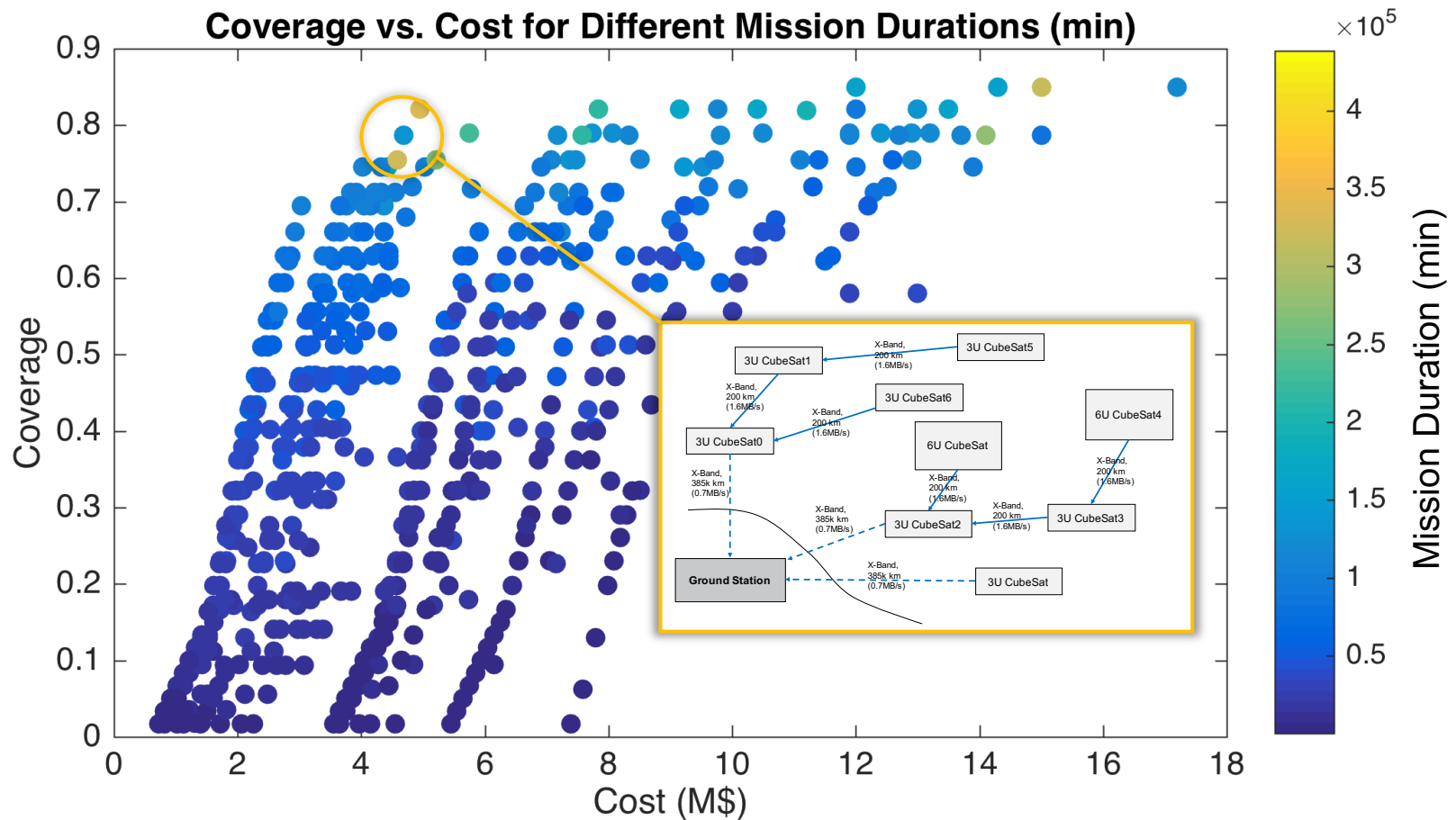
Open Source Technologies Used in Implementation

- Kigen Modeling Language
→ **Xtext**
- Representation of Domain
→ **Ecore / Eclipse EMF + OCL**
- Exploration Rules
→ **Henshin**
- Analysis Models / Fitness Functions
→ **Java**
- Optimization Using Genetic Algorithms
→ **MOMoT, MOEA**



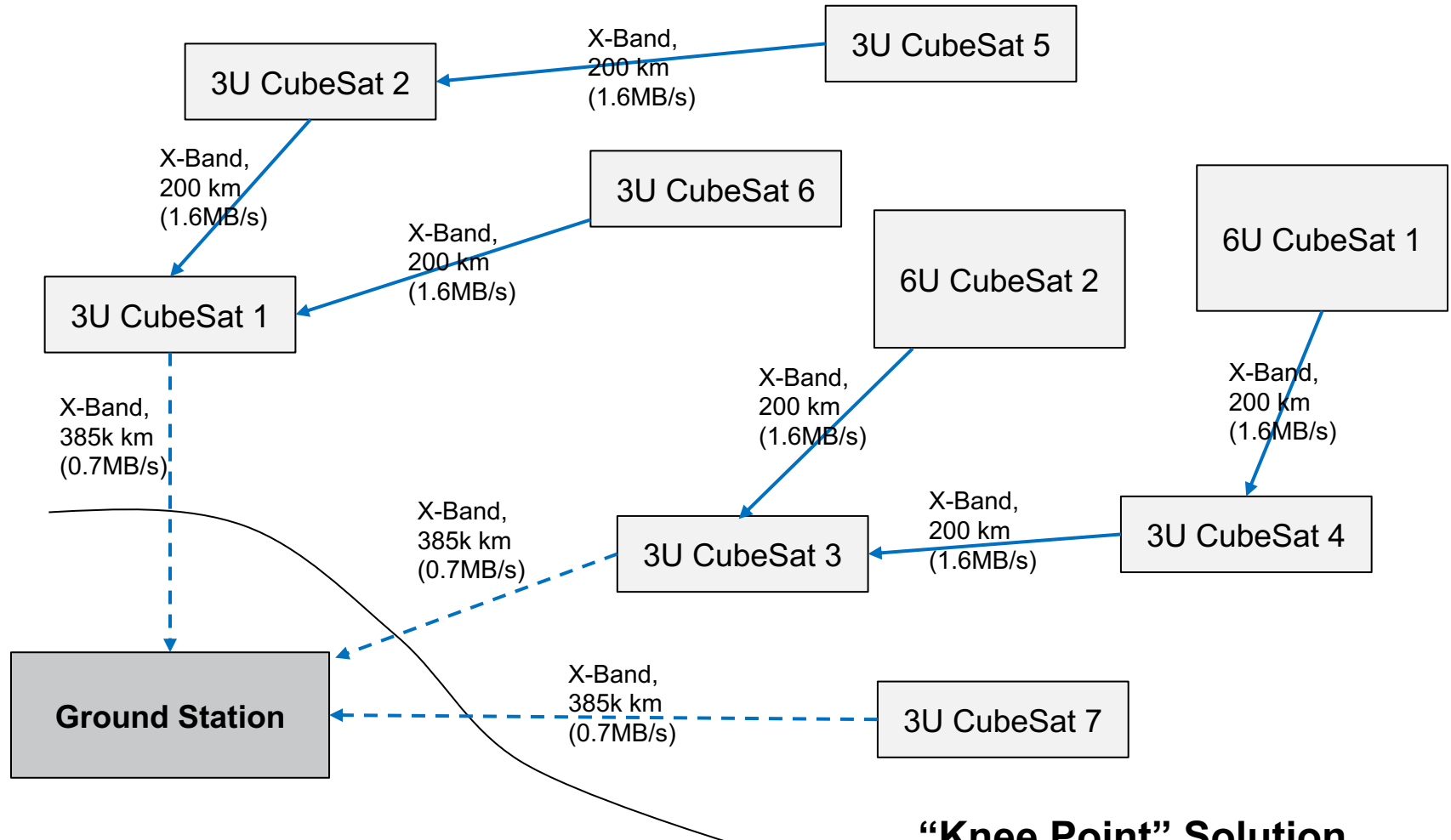
Results from Application to Case Study

Visualization of Trade Space



Results from Application to Case Study

Knee Point Solution



“Knee Point” Solution
\$4.7M, ~0.79 coverage (10h observation)

Conclusions

- Demonstrated the use of AI methods for supporting design & systems engineers in exploring a highly complex trade space
 - Able to generate diverse set of alternatives in a reasonable timeframe
 - Solutions determined based on abstract description of problem
 - Enables designers to focus on **analysis**, and consider more options
- The generated candidate solutions can [help spark creativity](#), but the method doesn't replace a team of engineers
 - Meant to **support** designers in the creative process
 - Analyzing results of optimizer can reveal missing information if results don't seem sensible
- [Good performance](#) for problems with limited scope, but should investigate methods for automatically dividing into sub-problems



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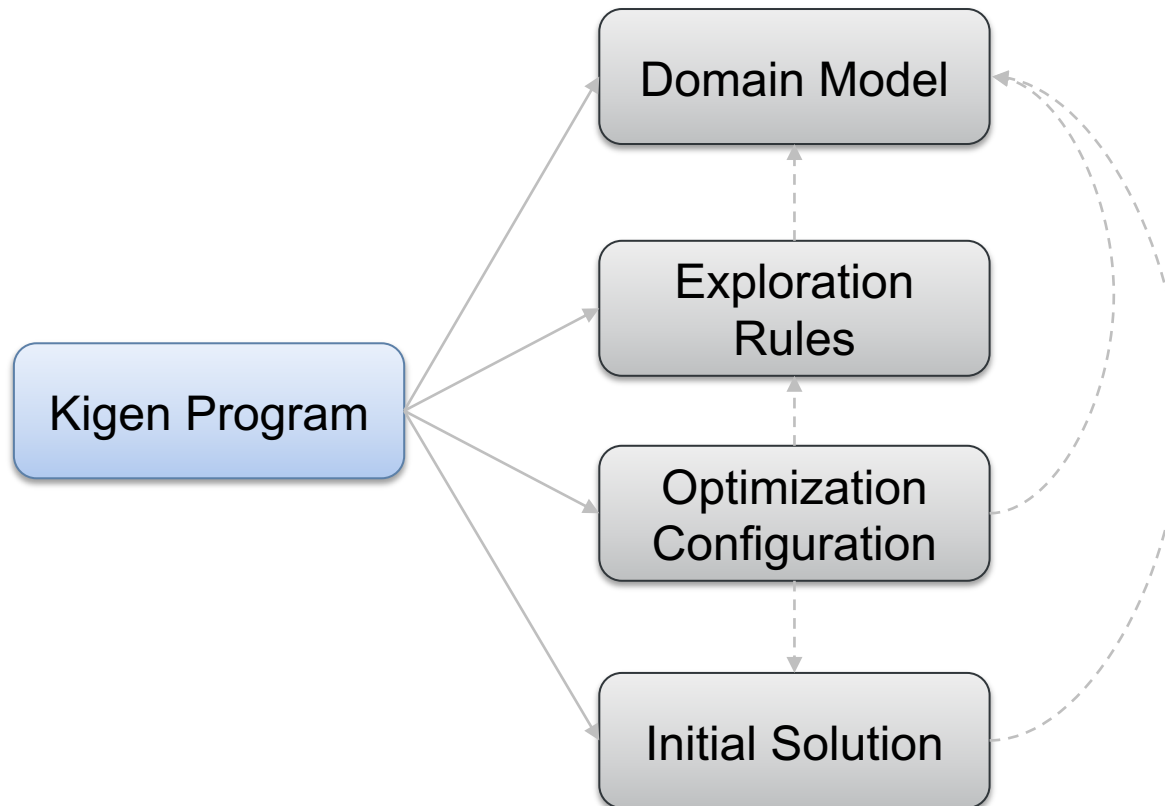
jpl.nasa.gov

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from publicly available sources and / or is fictitious.



Backup Slides

Derived Artifacts from a Kigen Program

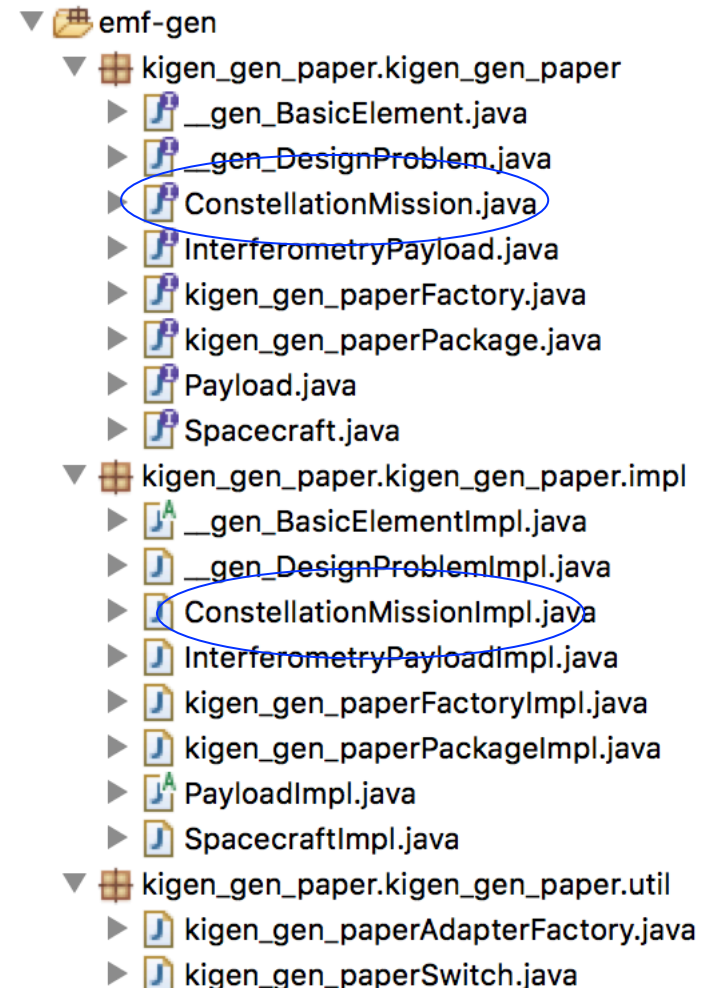


Generated Code

- A Java class is created for each “type” – code generation is based on GenModel mechanism in EMF
- These classes can be directly used in externally defined simulation models

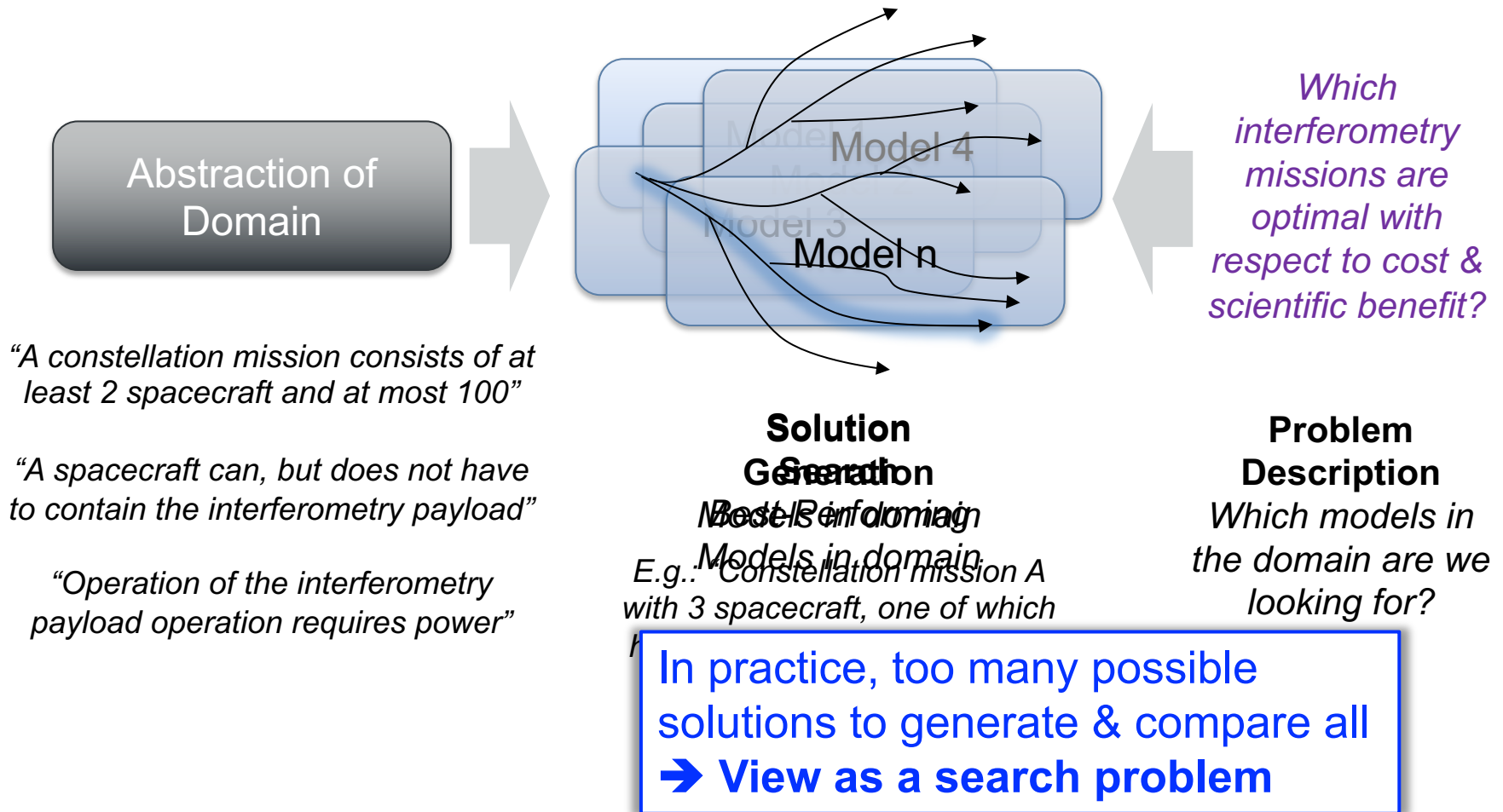
```
// KigenML Program
type ConstellationMission {
    ...
    fun someComplexSimulation =
        jvmcall mysim.Sim.simulate(this)
    ...
}

// Java Program
public class Sim {
    public static double simulate(
        ConstellationMission theMission) {
        return 0.0;
    }
}
```



Finding Models in Domain

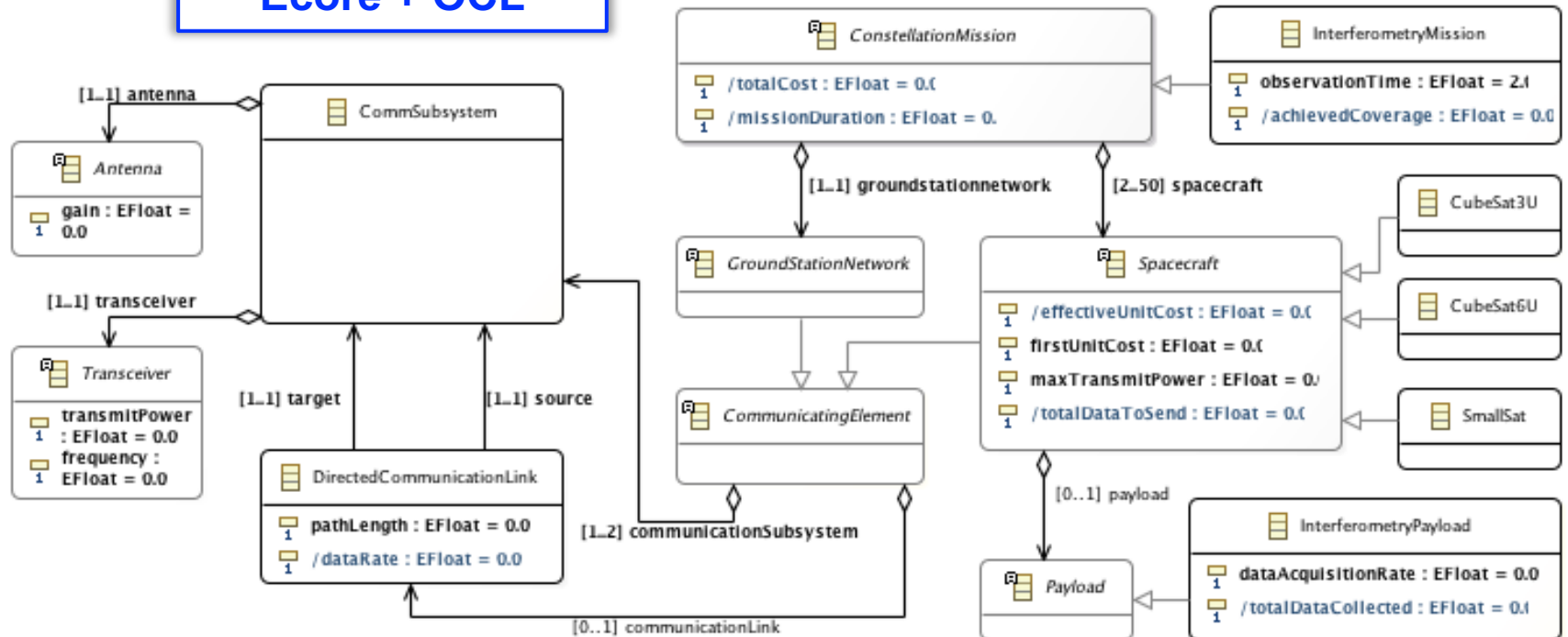
Mechanized Exploration



Application to Case Study

Representation of Domain (Excerpt)

Domain model in Ecore + OCL



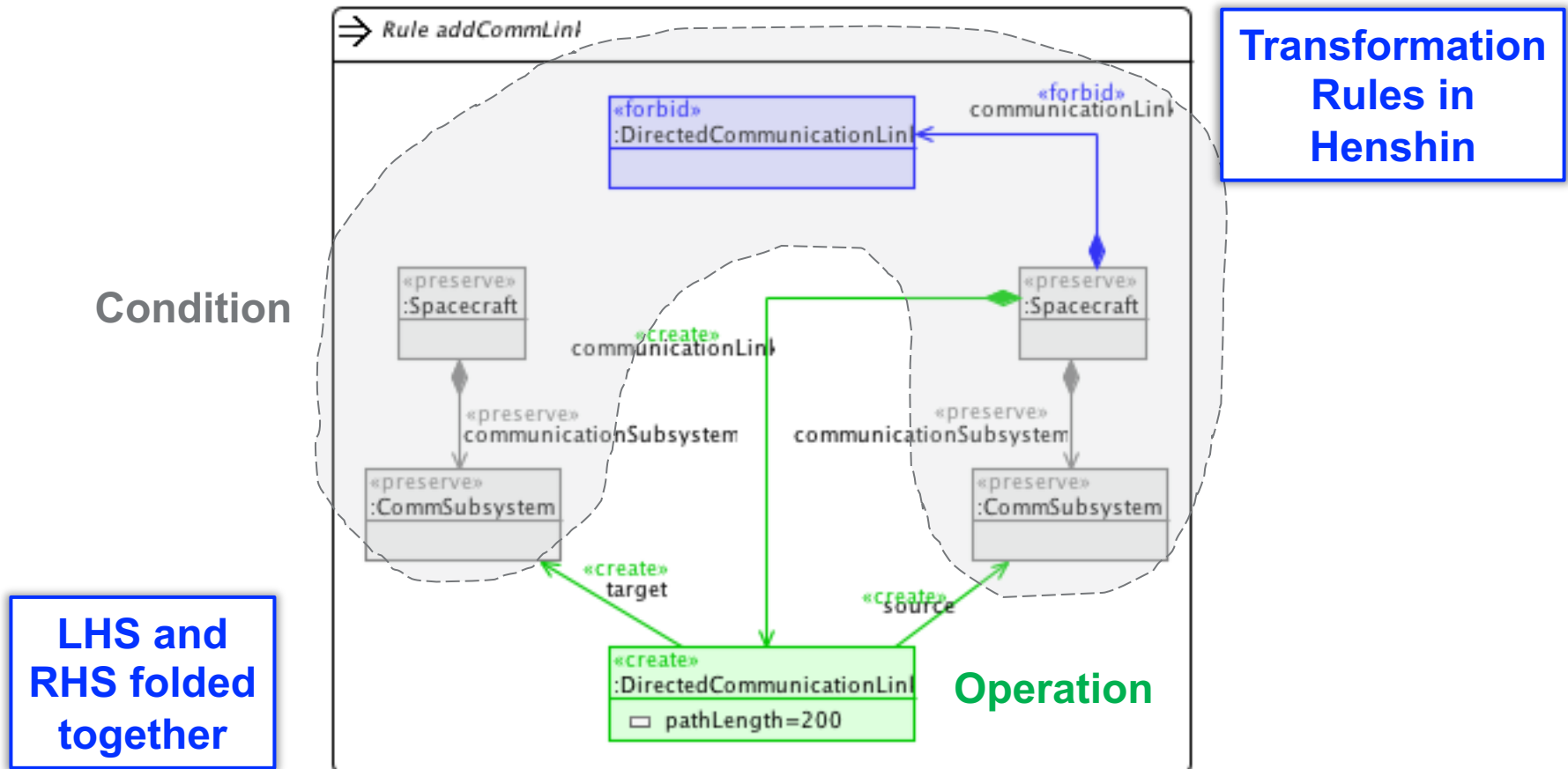
20 concepts, 9 associations, 15 attributes / parameters

> 48¹⁰ possible models

Too many for
exhaustive search

Application to Case Study

Transformation Rule Example (Henshin Syntax): Add Comm. Link

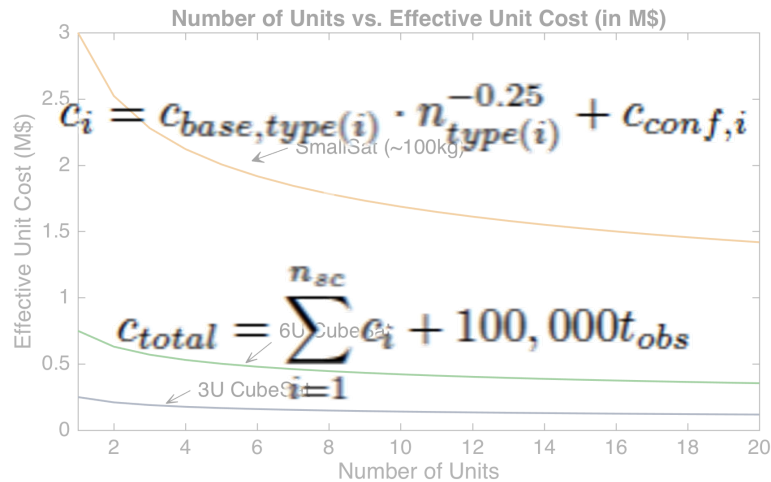


In Prose: “Find 2 distinct spacecraft instances, and add a communication link between them”

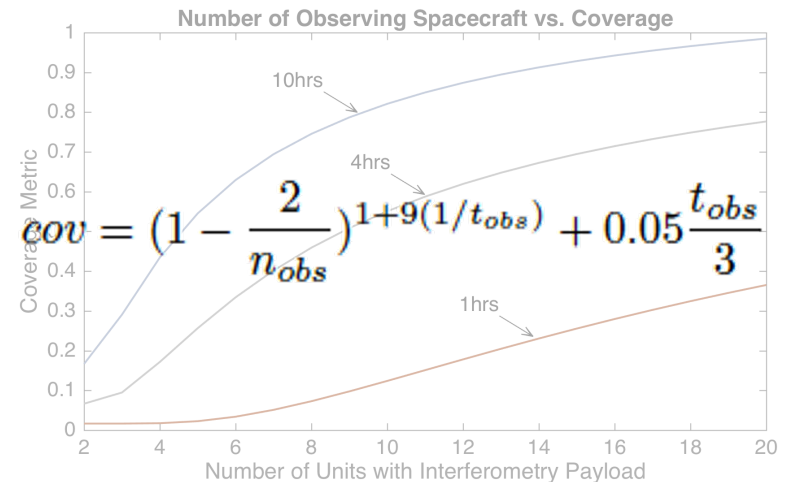
Application to Case Study

Analysis Model Definitions

Cost Model



Science Model



Data Management Model

- Data out = own science + data in
- Simplified operations

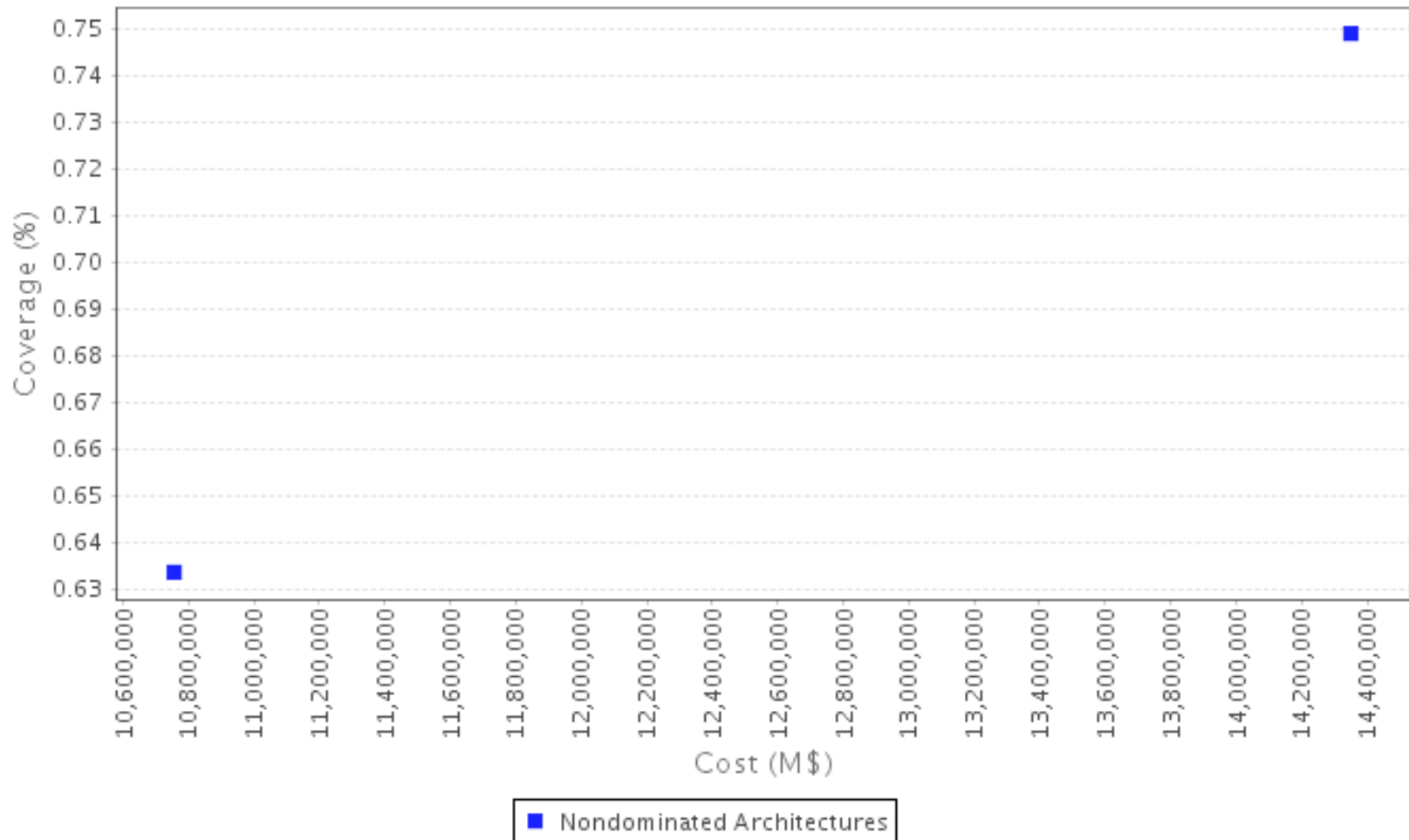
$$d_{out,i} = d_{science,i} + \sum_{j=1}^{n_{sc}} d_{in,j}$$

Communication / Link Model

Transmitter Configuration	200 km	385k km
UHF, 3 W, 1 dBi	5 Mbps	-
X-Band, 5 W, 10 dBi	1.6 Mbps	0.7 Mbps
Ka-Band, 15 W, 25 dBi	220 Mbps	80 Mbps

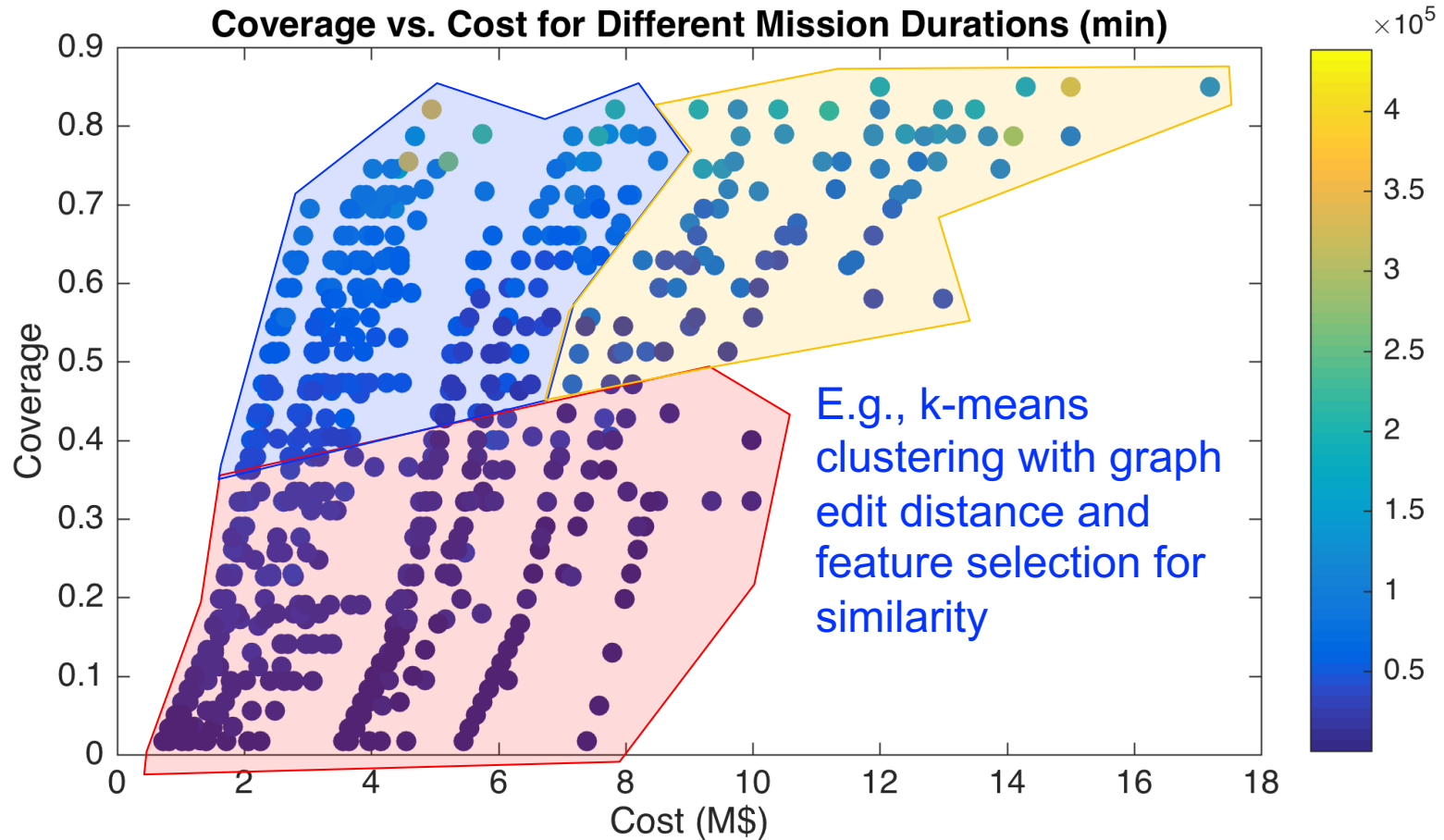
Evolution of Population (Algorithm: NSGA-II)

Achieved Coverage (%) vs. Cost (M\$) vs. Mission Duration (s)



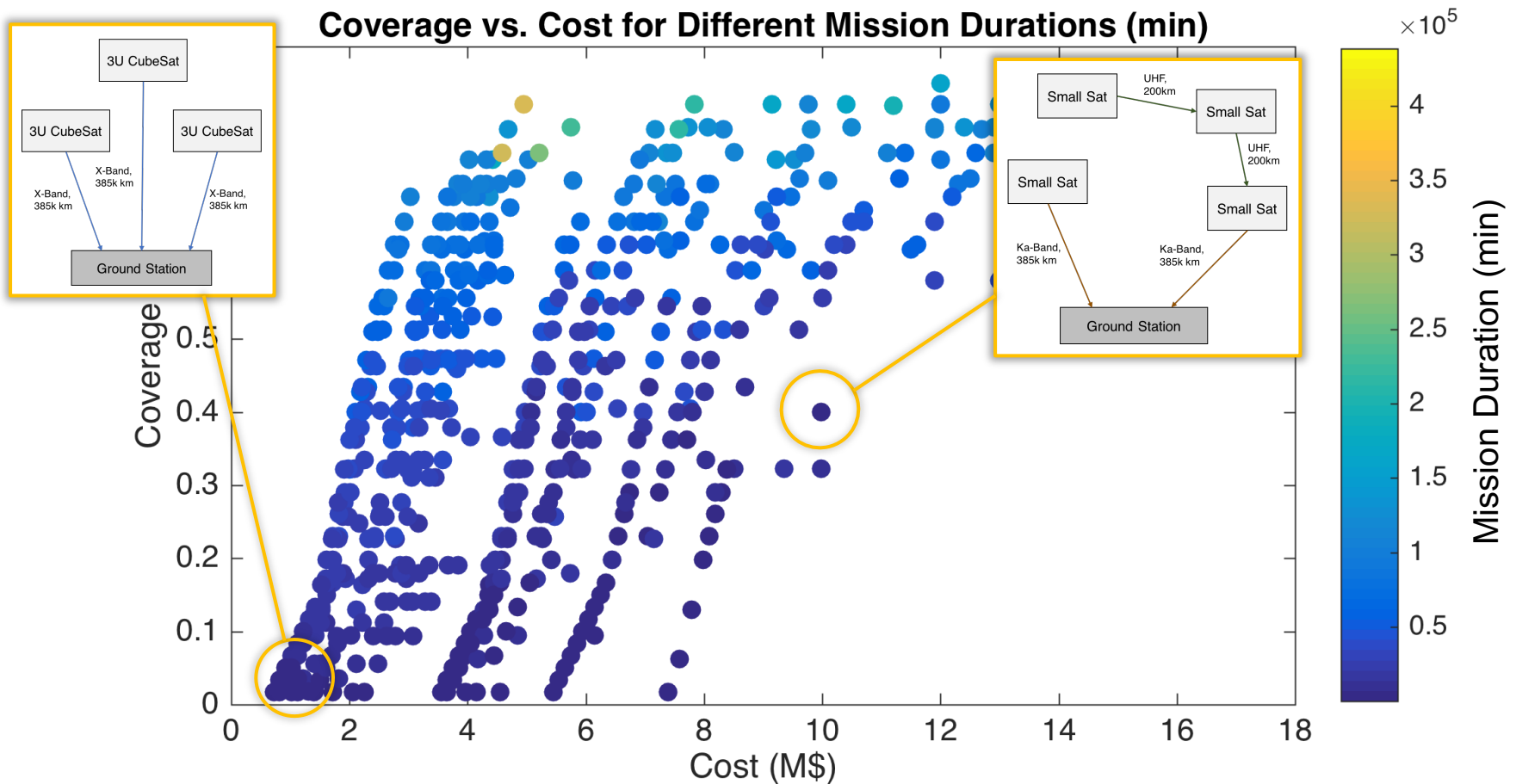
Additional Analysis

Clustering of Solutions



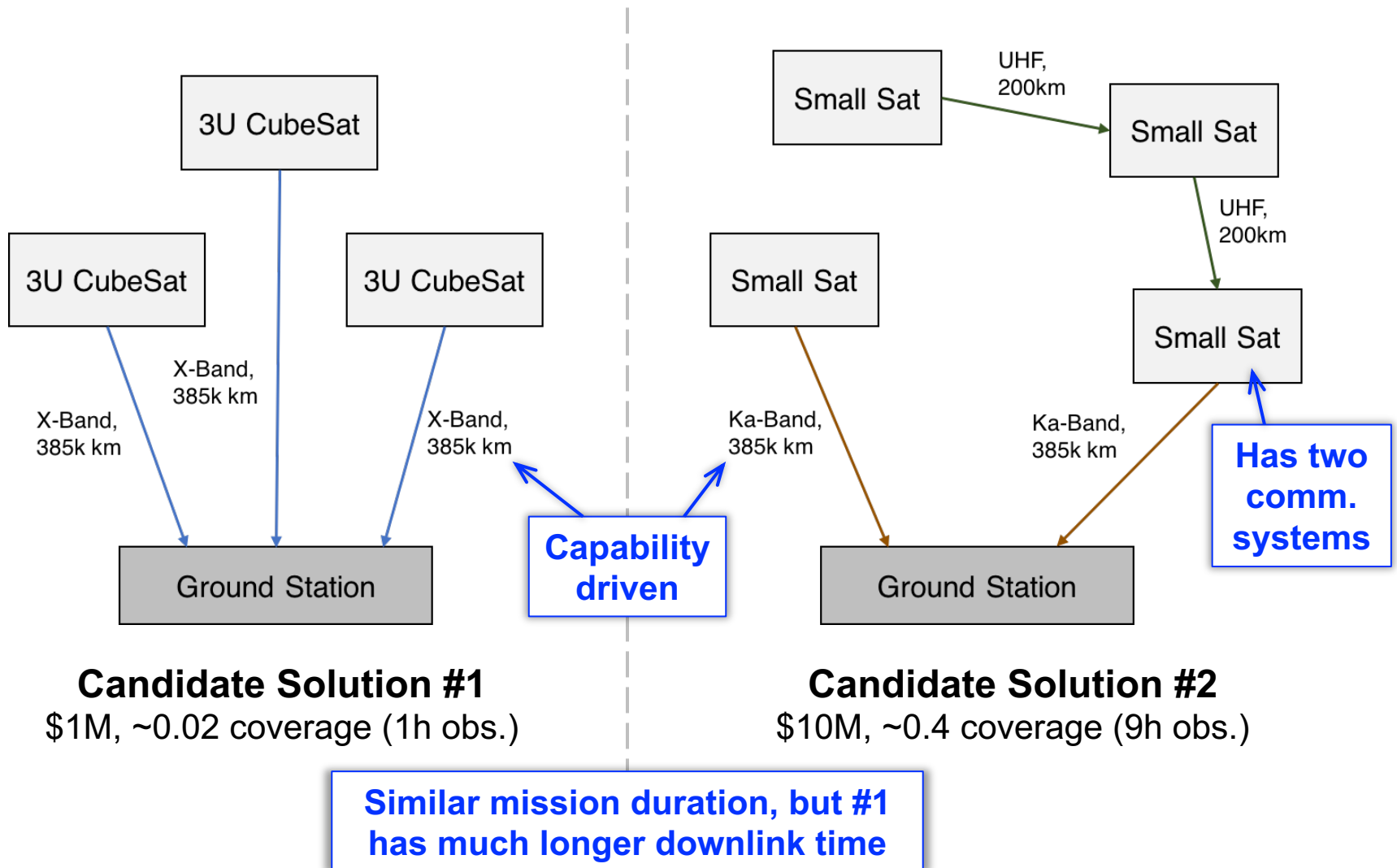
Results from Application to Case Study

Visualization of Trade Space



Results from Application to Case Study

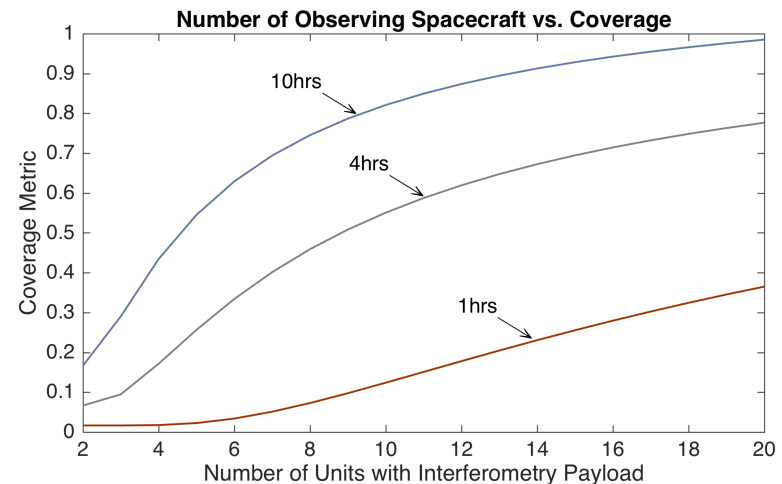
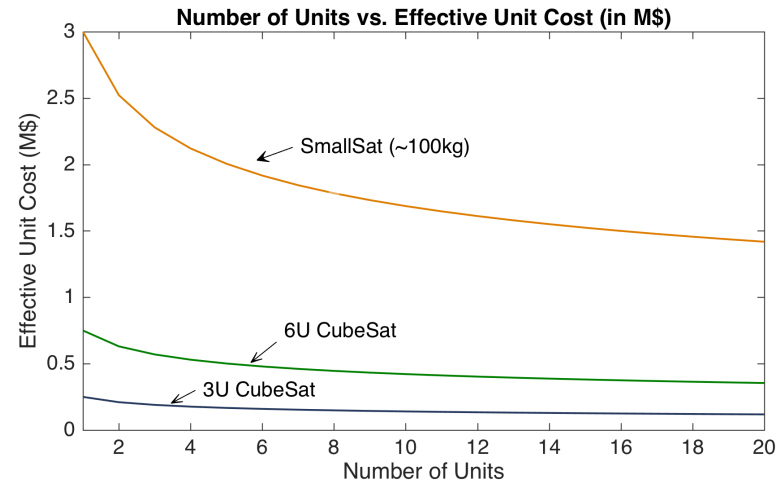
Examples of Pareto-Optimal (Nondominated) Solutions



Application to Case Study

- Three objectives:
 - Minimize **cost**
 - Maximize **coverage** (measure of scientific benefit)
 - Minimize **mission time**
- Typical link budget for data rates
- Data collection & transfer model
- Abstracted away orbit design through coverage model
- Experiment setup:
 - 16 transformation rules
 - 180 variables per individual
 - NSGA-II with population size 1000, and 100 generations
 - 30 runs*

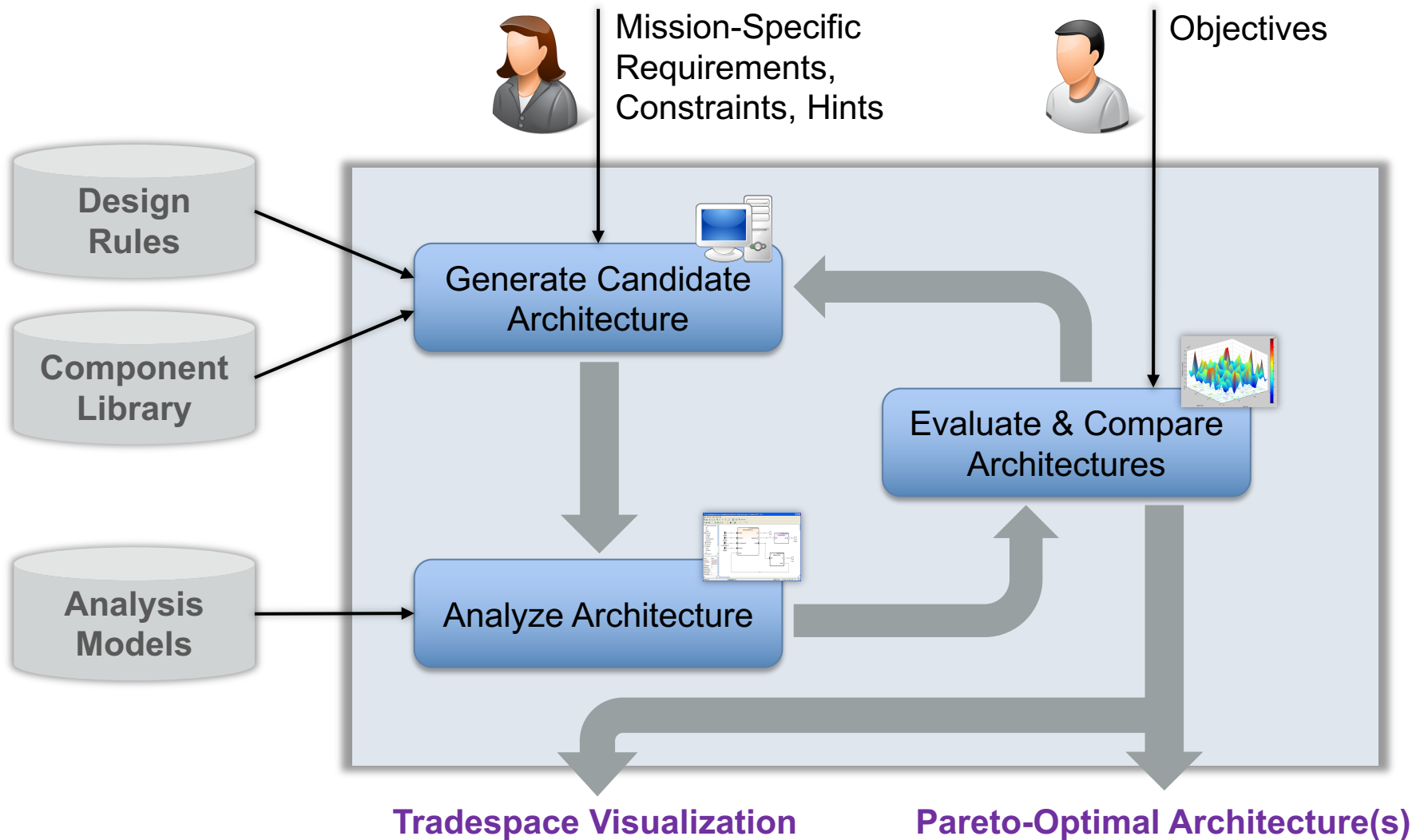
* 8 core Intel i7 @ 2.7Ghz, 16GB DDR3 RAM



*Fictitious cost model (top)
and coverage model (bottom)*

Framework

CDS for Mission Architecture Design



Application to Case Study

Link Calculations

- Derived from standard link budget, assuming above average noise due to expected interference from Moon

Table 1. Computed communication rates. 385k km case assumes 72 dBi receive antenna gain for X-band, and 85 dBi for Ka-band (similar to DSN).

Transmitter Configuration	200 km	385k km
UHF, 3 W, 1 dBi	5 Mbps	-
X-Band, 5 W, 10 dBi	1.6 Mbps	0.7 Mbps
Ka-Band, 15 W, 25 dBi	220 Mbps	80 Mbps

Application to Case Study

Cost Calculations

- Cost per spacecraft calculation incorporates a learning curve
- Assuming \$ 100,000 per hour of observation to estimate observation and data processing cost

$$c_i = c_{base,type(i)} \cdot n_{type(i)}^{-0.25} + c_{conf,i} \quad (5)$$

$$c_{total} = \sum_{i=1}^{n_{sc}} c_i + 100,000 t_{obs} \quad (6)$$

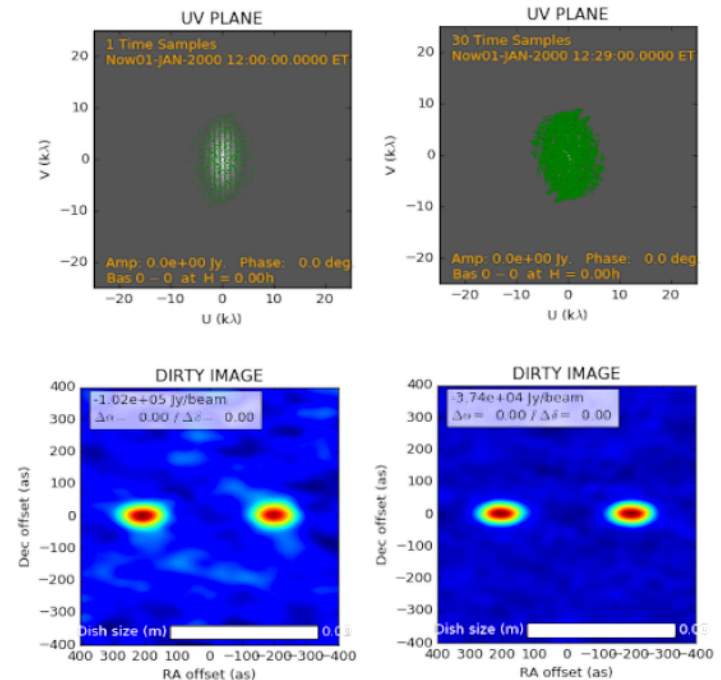
Application to Case Study

Coverage

- Simple coverage calculation

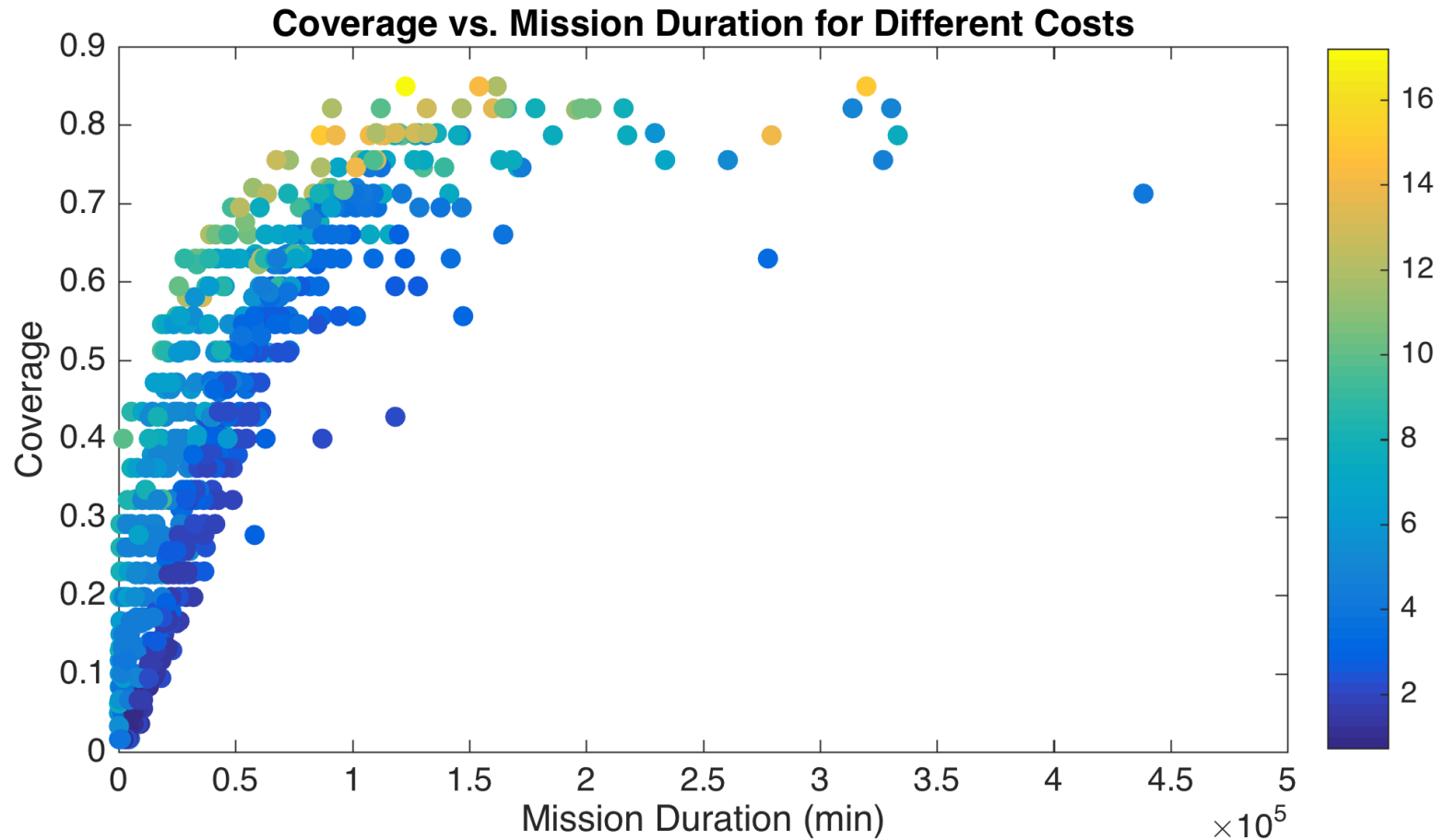
$$cov = \left(1 - \frac{2}{n_{obs}}\right)^{1+9(1/t_{obs})} + 0.05 \frac{t_{obs}}{3} \quad (1)$$

- Surrogate model that reflects trends observed from more sophisticated telescope array simulation performed by Alexander Hegedus (<https://github.com/alexhege/Orbital-APSYNSIM/>)



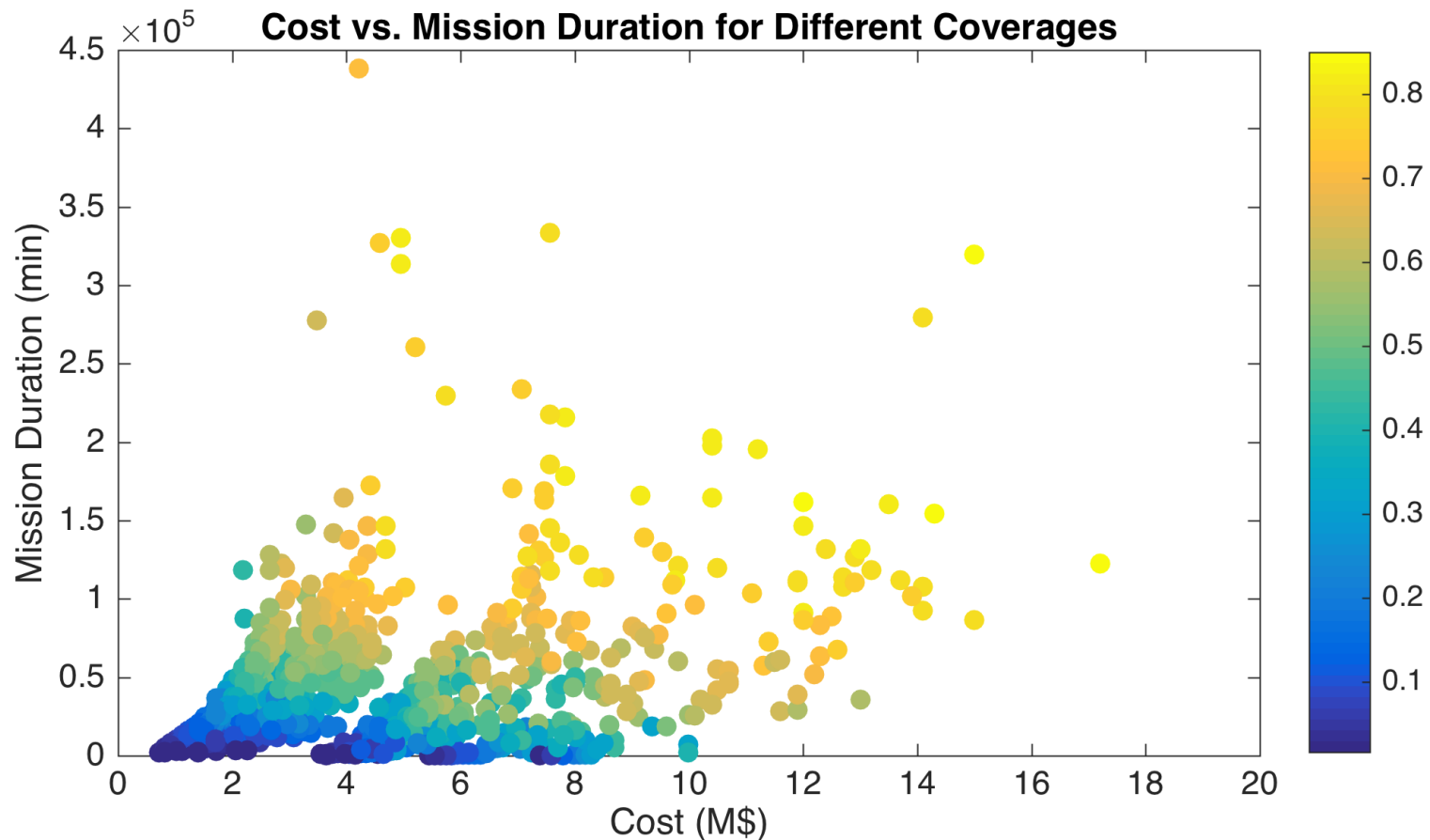
Results from Application to Case Study

Coverage vs. Mission Duration



Results from Application to Case Study

Cost vs. Mission Duration



Motivating Case: Mars Cave Exploration

Select Network-, Vehicle- and Operations-Level Trades

Option 1: smaller number of nodes, higher risk of comm. failure

Cheap, risky

Maximize distance into cave
(at least 100m)

Minimize cost, risk

Science Tasks

1. Map out cave
→ LiDAR? Stereo Imaging?

2. Measure environmental properties
→ temp, rad, ...

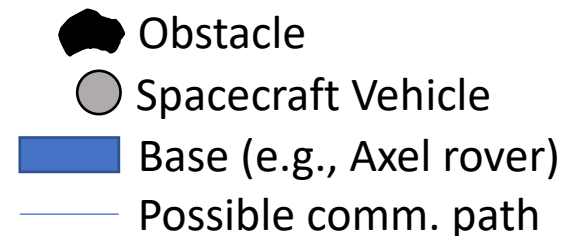
3. Secondary measurements

Maximize science return

Option 2: larger number of nodes, higher connectivity (smaller risk of network failure)

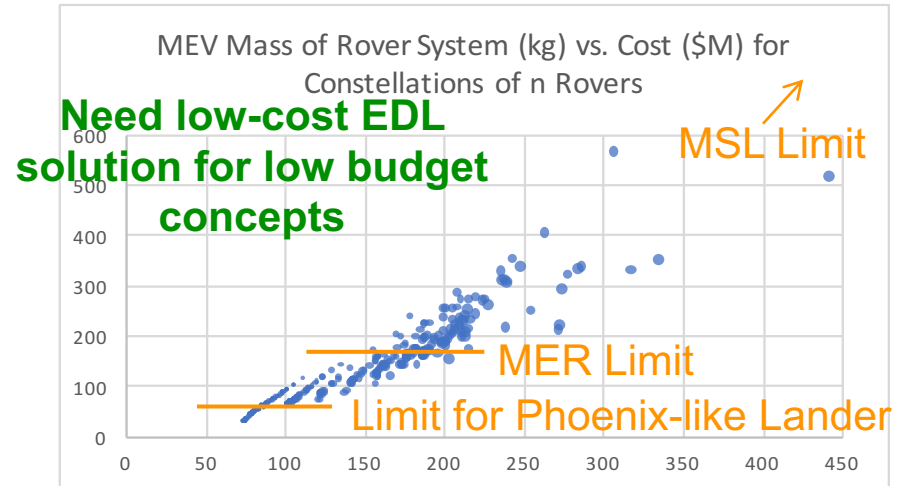
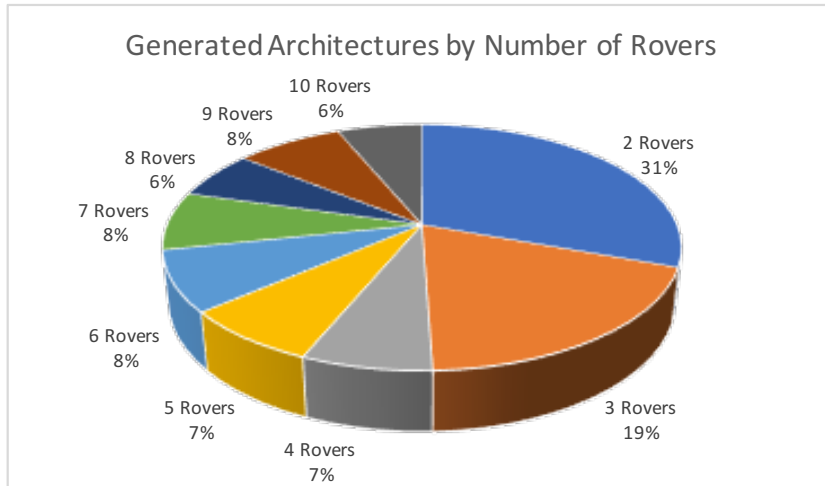
Expensive, robust

Small spacecraft ~ limited capabilities
→ **How many rovers? Which assets are equipped with which instruments?**

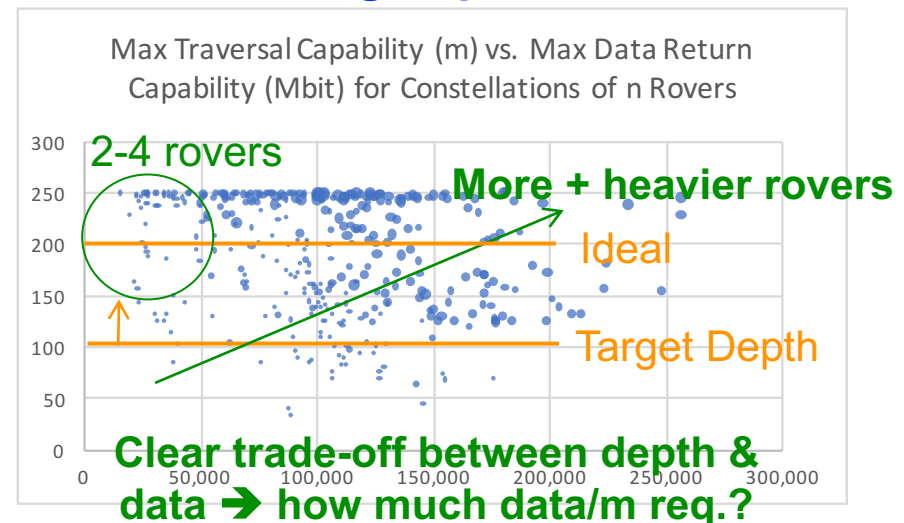
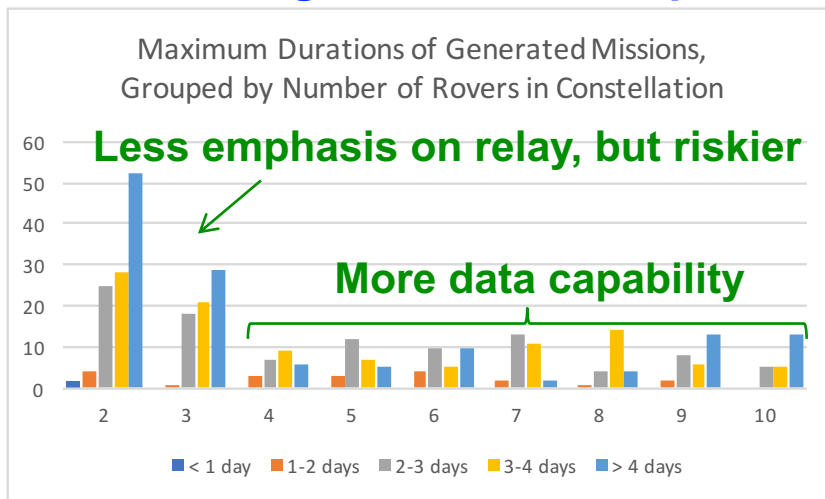


Analysis & Visualization of Optimizer Results

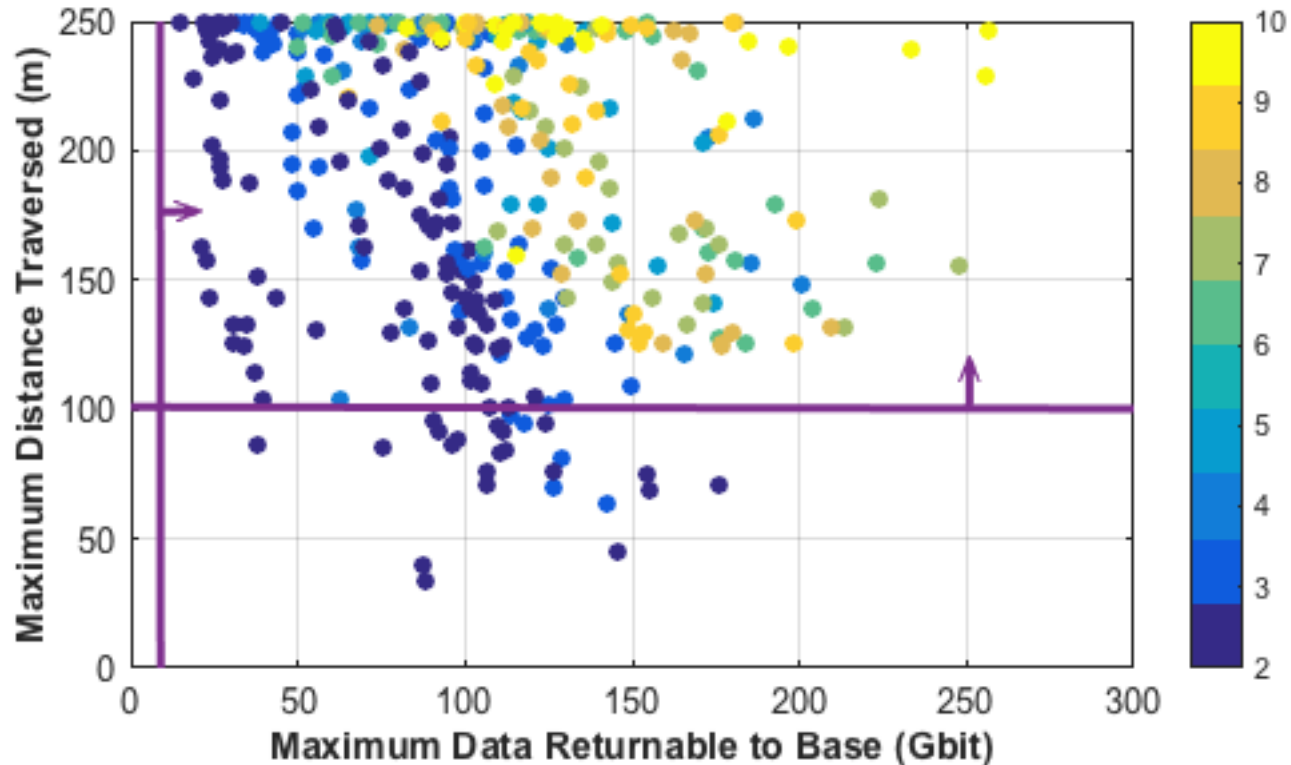
Analysis of Feasible Solutions (l) & Visualization of Trade Space (r)



→ Integrated vehicle, operations & network design optimization

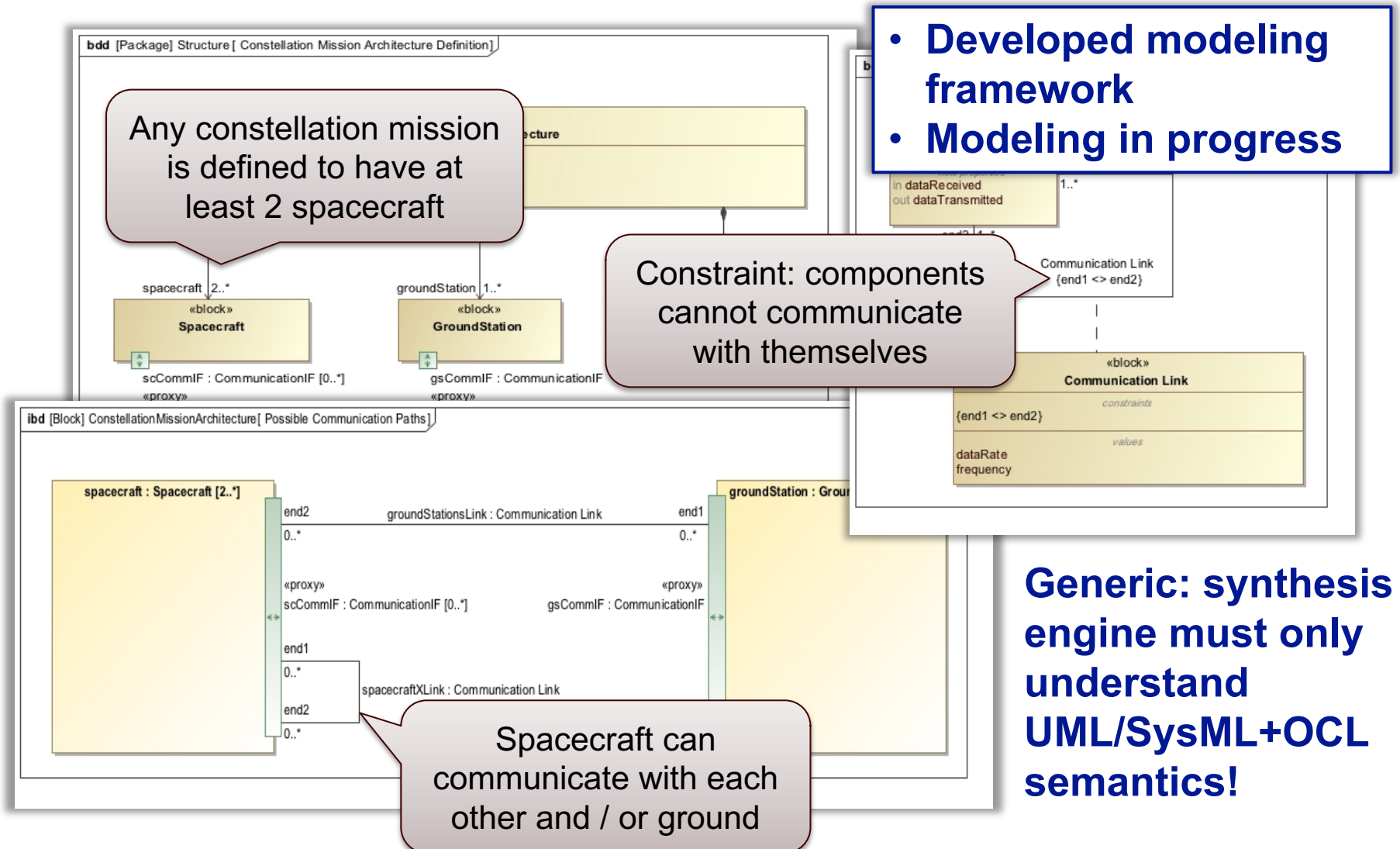


Results from Mars Cave Exploration Mission Case Study



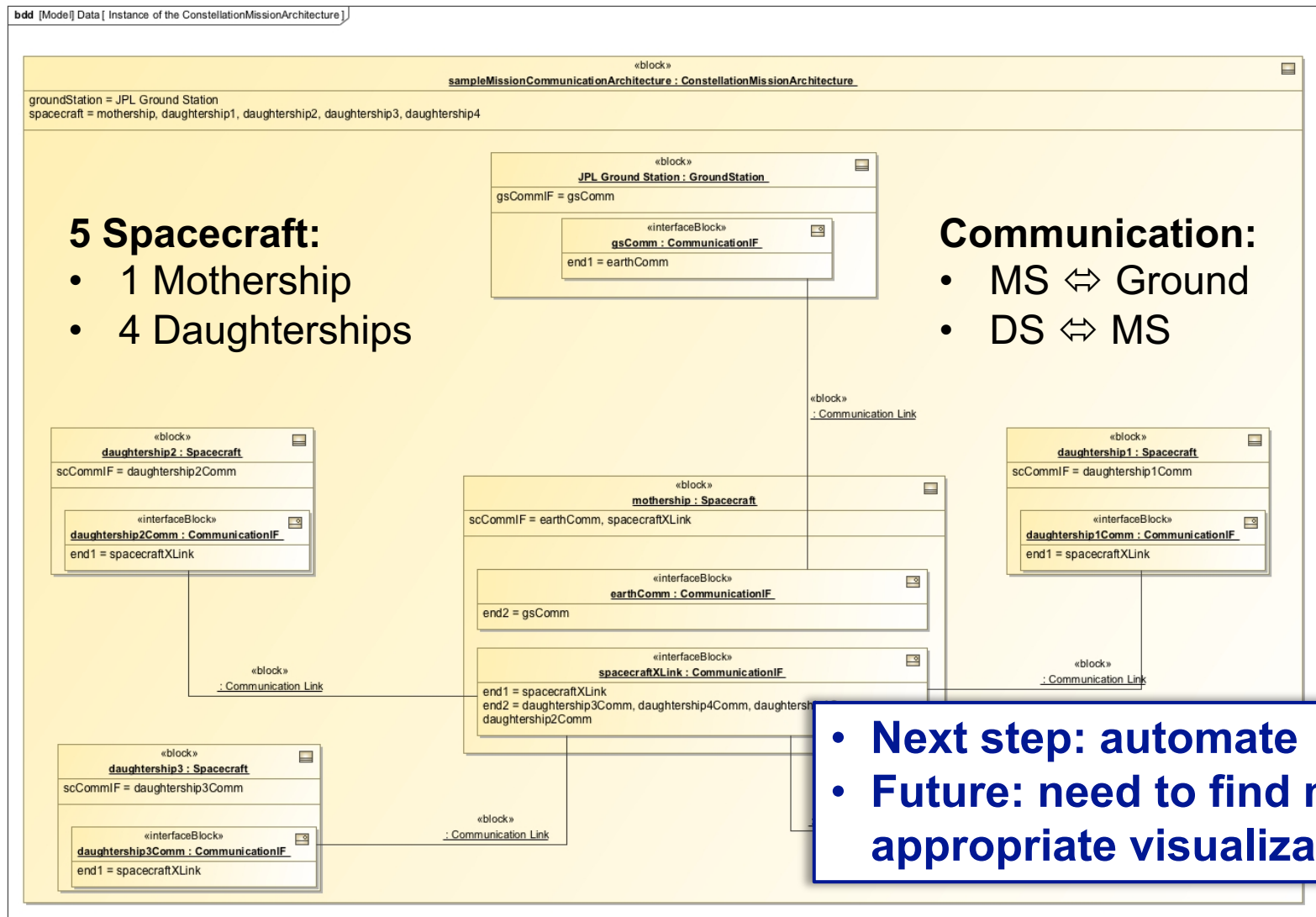
Capturing Design Options & Rules Formally

Optimized Integrated Network Constellation Design



Instance Creation: Example Partial Instance

Optimized Integrated Network Constellation Design

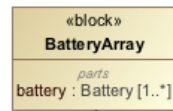
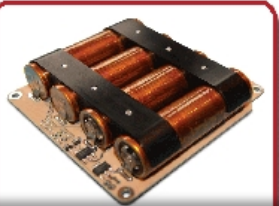


Component & Analysis Library

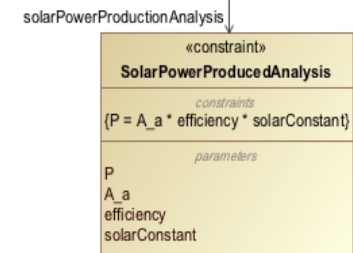
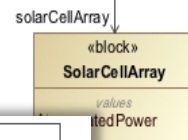
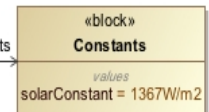
Optimized Integrated Network Constellation Design

Results from team, DARPA F6 and TeamXc leveraged for component & analysis library

“Templates” for instance creation



bdd [Package] S



Can represent arbitrary analysis model

Analysis context separates analysis & design concerns

Content / Outline

- Context / problem statement 2min [p]
- Challenge 1min [r]
- Approach
 - Rule-based exploration, meta-model + rules, chain of rules 2min [a]
 - KigenML to specify, include objectives 1min [n]
 - Optimization mechanism(s) (+ mention post-processing) 1.5min [a]
- Results
 - Interferometer 1min [r]
 - [Cave] 1min [r]
 - (Just mention clustering)
 - (What was surprising?)
- Conclusions 1min [n]